

Constraining the astrophysics and cosmology from 21cm with the SKA

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Citation Report

#	ARTICLE	IF	CITATIONS
1	Cosmological parameter estimation from large-scale structure deep learning. <i>Science China: Physics, Mechanics and Astronomy</i> , 2020, 63, 1.	2.0	24
2	A new way to constrain the densities of intragroup medium in groups of galaxies with convolutional neural networks. <i>Monthly Notices of the Royal Astronomical Society</i> , 2020, 497, 5090-5102.	1.6	3
3	Deep learning for intensity mapping observations: component extraction. <i>Monthly Notices of the Royal Astronomical Society: Letters</i> , 2020, 496, L54-L58.	1.2	15
4	Investigating X-Ray Sources during the Epoch of Reionization with the 21 cm Signal. <i>Astrophysical Journal</i> , 2021, 912, 143.	1.6	12
5	Cosmic Velocity Field Reconstruction Using AI. <i>Astrophysical Journal</i> , 2021, 913, 2.	1.6	11
6	Deep learning approach for identification of H&scii regions during reionization in 21-cm observations. <i>Monthly Notices of the Royal Astronomical Society</i> , 2021, 505, 3982-3997.	1.6	16
7	Predicting 21&scm-line map from Lyman-Î emitter distribution with generative adversarial networks. <i>Monthly Notices of the Royal Astronomical Society</i> , 2021, 506, 357-371.	1.6	4
8	Hlnet: Generating Neutral Hydrogen from Dark Matter with Neural Networks. <i>Astrophysical Journal</i> , 2021, 916, 42.	1.6	16
9	The CAMELS Project: Cosmology and Astrophysics with Machine-learning Simulations. <i>Astrophysical Journal</i> , 2021, 915, 71.	1.6	113
10	Removing Astrophysics in 21 cm Maps with Neural Networks. <i>Astrophysical Journal</i> , 2021, 907, 44.	1.6	27
11	Constraining the Reionization History using Bayesian Normalizing Flows. <i>Machine Learning: Science and Technology</i> , 2020, 1, 035014.	2.4	13
12	Inference from the 21 cm Signal. , 0, , .		0
13	Machine learning astrophysics from 21&cm lightcones: impact of network architectures and signal contamination. <i>Monthly Notices of the Royal Astronomical Society</i> , 2021, 509, 3852-3867.	1.6	22
14	Inferring astrophysics and dark matter properties from 21 cm tomography using deep learning. <i>Monthly Notices of the Royal Astronomical Society</i> , 2022, 511, 3446-3462.	1.6	4
15	Estimation of H II Bubble Size Distribution from 21 cm Power Spectrum with Artificial Neural Networks. <i>Research in Astronomy and Astrophysics</i> , 0, , .	0.7	2
16	Deep learning the astrometric signature of dark matter substructure. <i>Physical Review D</i> , 2021, 104, .	1.6	3
17	Simulation-based Inference of Reionization Parameters from 3D Tomographic 21 cm Light-cone Images. <i>Astrophysical Journal</i> , 2022, 926, 151.	1.6	27
18	Exploring the cosmic 21-cm signal from the epoch of reionization using the wavelet scattering transform. <i>Monthly Notices of the Royal Astronomical Society</i> , 2022, 513, 1719-1741.	1.6	10

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19	Cosmology with One Galaxy?. <i>Astrophysical Journal</i> , 2022, 929, 132.	1.6	10
20	Exploring the cosmic dawn and epoch of reionization with the 21-cm line. <i>Publication of the Astronomical Society of Japan</i> , 2023, 75, S1-S32.	1.0	2
21	Probing the Inflaton Potential with SKA. <i>SciPost Physics Core</i> , 2022, 5, .	0.9	1
22	Implicit Likelihood Inference of Reionization Parameters from the 21 cm Power Spectrum. <i>Astrophysical Journal</i> , 2022, 933, 236.	1.6	12
23	A bubble size distribution model for the Epoch of Reionization. <i>Astronomy and Astrophysics</i> , 2022, 667, A118.	2.1	2
24	HiFlow: Generating Diverse Hi Maps and Inferring Cosmology while Marginalizing over Astrophysics Using Normalizing Flows. <i>Astrophysical Journal</i> , 2022, 937, 83.	1.6	5
25	Learning Cosmology and Clustering with Cosmic Graphs. <i>Astrophysical Journal</i> , 2022, 937, 115.	1.6	17
26	Inpainting Hydrodynamical Maps with Deep Learning. <i>Astrophysical Journal</i> , 2022, 941, 132.	1.6	1
27	Detecting the non-Gaussianity of the 21-cm signal during reionization with the wavelet scattering transform. <i>Monthly Notices of the Royal Astronomical Society</i> , 2023, 519, 5288-5303.	1.6	4
28	Artificial neural networks for galaxy clustering: Learning from the two-point correlation function of BOSS galaxies. <i>Astronomy and Computing</i> , 2023, 42, 100692.	0.8	0
29	Test of artificial neural networks in likelihood-free cosmological constraints: A comparison of information maximizing neural networks and denoising autoencoder. <i>Physical Review D</i> , 2023, 107, .	1.6	1
30	The CAMELS Project: Public Data Release. <i>Astrophysical Journal, Supplement Series</i> , 2023, 265, 54.	3.0	14